

```
[1]: ##imports from libraries
import pandas as pd
import numpy as np
import time
import math
from sklearn.model_selection import train_test_split
import matplotlib.pyplot as plt
import resource
import psutil

[2]: ## Preprocessing of data
# Load data here:

data1 = pd.read_csv('CA1_1c_greenhouse_cleanedData_part1.csv')
data2 = pd.read_csv('CA1_1c_greenhouse_cleanedData_part2.csv')
data = pd.concat([data1,data2],ignore_index=True)

[3]: #data.describe()

[4]: ## Find the number of 'nan' in each column:
#data.isnull().sum()

[5]: data.shape

[5]: (2921, 5232)

[6]: ## Defien X and Y
X=data.iloc[:,0:5231]
Y=data.iloc[:,5231]

#x_mean=X.mean()
#x_std= X.std()
#X=(X-x_mean)/x_std

# Split train and test data here: (X_train, Y_train, X_test, Y_test)
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2,
↪random_state=42)

[7]: X_train.shape

[7]: (2336, 5231)

[8]: Y_train.shape

[8]: (2336,)

[9]: ## Logistic ridge regression with different optimizers
# cost function
```

```

# N: number of samples, D: number of features

def cost(X, y, w, lambda_):
    N, D = X.shape
    value = 0
    for i in range(N):
        Z = -1 * y[i] * np.matmul(w.T , X[i,:].reshape(D,1))
        value += np.log(1+np.exp(Z))
    norm_w = np.linalg.norm(w)
    c = lambda_ * norm_w ** 2
    return value/N + c

```

```

[10]: # cost function and gradient calculation
def grad(X, y, w, lambda_):
    # Calculate the gradient here
    N, D = X.shape
    gradient = np.zeros((D, 1))
    y = np.array(y).reshape(-1, 1) # convert y to numpy array and reshape
    for i in range(N):
        Z = y[i] * np.matmul(w.T , (X[i, :]).reshape(D, 1)) # use X[i, :]
        ↪ instead of X[:, i]
        exp_Z = np.exp(Z)
        gradient += (-y[i] / (1 + exp_Z)) * (X[i, :]).reshape(D, 1)
    c = 2 * lambda_ * w # add regularization term
    return gradient/N + c

```

```

[15]: # ----- Complete the blank definitions:
        ↪ -----
def solver(x,y, w, alpha, num_iters , lambda_ , epsilon , optimizer, mem):

    if (optimizer == "GD") :
        for i in range(num_iters):
            # update the parameter w for GD here:
            grad_w = grad(x, y, w, lambda_)
            w -= alpha * grad_w
            if (i%10==0) and (mem):
                usage=resource.getrusage(resource.RUSAGE_SELF)
                print("mem for GD (MB):", (usage[2]*resource.getpagesize())/
                ↪ 1000000.0)
            if (np.linalg.norm(grad_w) <= epsilon):
                break

    elif (optimizer == "SGD"):
        N, D = x.shape
        batch = 1
        for i in range(num_iters):
            # Complete SGD here:

```

```

        zeta = np.random.randint(0, N, batch)
        grad_w = grad(x[zeta,:].reshape(batch, x.shape[1]) , y[zeta].
→reshape(batch,1), w, lambda_)
        w -= alpha * grad_w
        if (i%10==0) and (mem):
            usage=resource.getrusage(resource.RUSAGE_SELF)
            print("mem for SGD (MB):", (usage[2]*resource.getpagesize())/
→1000000.0)
        if (np.linalg.norm(grad_w) <= epsilon):
            break

elif (optimizer == "SVRG"):
    T = 10
    K = math.floor(num_iters/T)
    #Z = np.matmul(x,np.diagflat(y))
    N, D = x.shape
    w_tilde = w
    for k in range(K):
        #Compute all gradients and store
        grad_avg = grad(x, y, w, lambda_)
        for t in range(T):
            # Complete SVRG here:
            zeta = np.random.randint(N)
            x_t = x[zeta].reshape(-1, D) # Reshape X to be 2D
            y_t = y[zeta]
            grad_w_tilde = grad(x_t, y_t , w_tilde, lambda_)
            grad_w = grad(x_t, y_t, w, lambda_)
            w_tilde -= alpha * (grad_w_tilde - grad_w + grad_avg)
        w = w_tilde
        #if (i%10==0) and (mem):
        #    usage=resource.getrusage(resource.RUSAGE_SELF)
        #    print("mem for SVRG (MB):", (usage[2]*resource.getpagesize())/
→1000000.0)
        if (np.linalg.norm(grad_avg) <= epsilon):
            break

elif (optimizer == "SAG"):
    N, D = x.shape
    grad_avg = grad(x, y, w, lambda_)
    grad_memory = np.matmul (np.ones((N,1)), grad_avg.T)
    for i in range(num_iters):
        # Compute average gradient
        zeta = np.random.randint(N)
        grad_zeta = grad(x[zeta, :].reshape(-1,D), y[zeta], w, lambda_)
        grad_memory[zeta, :] = grad_zeta.T

```

```

        grad_avg = np.mean(grad_memory, axis=0).reshape(D,1)
        w = w - alpha * grad_avg
        if (i%10==0) and (mem):
            usage=resource.getrusage(resource.RUSAGE_SELF)
            print("mem for SAG (MB):", (usage[2]*resource.getpagesize())/
↪1000000.0)
            if (np.linalg.norm(grad_avg) <= epsilon):
                break
    return w

```

```

[19]: ## Define solvers: GD, SGD, SVRG and SAG.
      # Setting the values here:

      alpha = 0.1 # change the value
      num_iters = 50 #[1:5:30] # change the value
      lambda_ = 0.1 #change the value
      epsilon = 1e-6 #change the value

```

```

[20]: y = np.array(Y_train.iloc[0:2336])
      x = np.array(X_train.iloc[0:2336,:])
      N,D = x.shape
      w_prime = np.random.rand(D,1)*0.01 # Initialization of w

```

```

[21]: ## Executing the iterations and plot the cost function here:
      ti= np.zeros((num_iters,4))
      cost_ = np.zeros((num_iters,4))
      for i in range(num_iters):
          print(".....",i,".....")
          #-----GD-----
          start = time.time()
          gde = solver(x, y, w_prime, alpha, num_iters=i, lambda_=lambda_,
↪epsilon=epsilon, optimizer = "GD", mem=False)
          end = time.time()
          cost_[i,0] = cost(x,y,gde,lambda_)
          ti[i,0] = end-start

          #-----SGD-----
          #Complete for SGD solver here :
          start = time.time()
          sgde = solver(x, y, w_prime, alpha, num_iters=i, lambda_=lambda_,
↪epsilon=epsilon, optimizer = "SGD", mem=False)
          end = time.time()

          cost_[i,1] = cost(x,y,sgde,lambda_)

          ti[i,1] = end-start

```

```

#-----SVRG-----
#Complete for SVRG solver here :
start = time.time()
svrge = solver(x,y, w_prime, alpha, num_iters=i, lambda_=lambda_,
→epsilon=epsilon, optimizer = "SVRG",mem=False)
end = time.time()

cost_[i,2] = cost(x,y,svrge,lambda_)

ti[i,2] = end-start

#-----SAG-----
#Complete for SAG solver here :
start = time.time()
sage = solver(x, y, w_prime, alpha, num_iters=i, lambda_=lambda_,
→epsilon=epsilon, optimizer = "SAG",mem=False)

end = time.time()

cost_[i,3] = cost(x,y,sage,lambda_)

ti[i,3] = end-start

## Pl the results:

10 = plt.plot(cost_[:,0],color="red")
# complete other plots here:
11 = plt.plot(cost_[:,1],color="blue")
12 = plt.plot(cost_[:,2],color="green")
13 = plt.plot(cost_[:,3],color="black")

plt.xlabel("Number of Iteration")
plt.ylabel("Cost")
plt.legend(['GD', 'SGD', 'SVRG', 'SAG'])
plt.show()

10 = plt.plot(ti[:,0],color="red")
# complete other plots here:
11 = plt.plot(ti[:,1],color="blue")
12 = plt.plot(ti[:,2],color="green")
13 = plt.plot(ti[:,3],color="black")

plt.xlabel("Number of Iteration")
plt.ylabel("Time (sec)")

```

```
plt.legend(['GD', 'SGD', 'SVRG', 'SAG'])
plt.show()
```

```
... 0 ...
```

```
/var/folders/l9/nl6ry63s0m3_35qt7mrjv0fh0000gn/T/ipykernel_15949/4027734134.py:9
: RuntimeWarning: overflow encountered in exp
  exp_Z = np.exp(Z)
```

```
... 1 ...
```

```
... 2 ...
```

```
... 3 ...
```

```
... 4 ...
```

```
... 5 ...
```

```
... 6 ...
```

```
... 7 ...
```

```
... 8 ...
```

```
... 9 ...
```

```
... 10 ...
```

```
... 11 ...
```

```
... 12 ...
```

```
... 13 ...
```

```
... 14 ...
```

```
... 15 ...
```

```
... 16 ...
```

```
... 17 ...
```

```
... 18 ...
```

```
... 19 ...
```

```
... 20 ...
```

```
... 21 ...
```

```
... 22 ...
```

```
... 23 ...
```

```
... 24 ...
```

```
... 25 ...
```

```
... 26 ...
```

```
... 27 ...
```

```
... 28 ...
```

```
... 29 ...
```

```
... 30 ...
```

```
... 31 ...
```

```
... 32 ...
```

```
... 33 ...
```

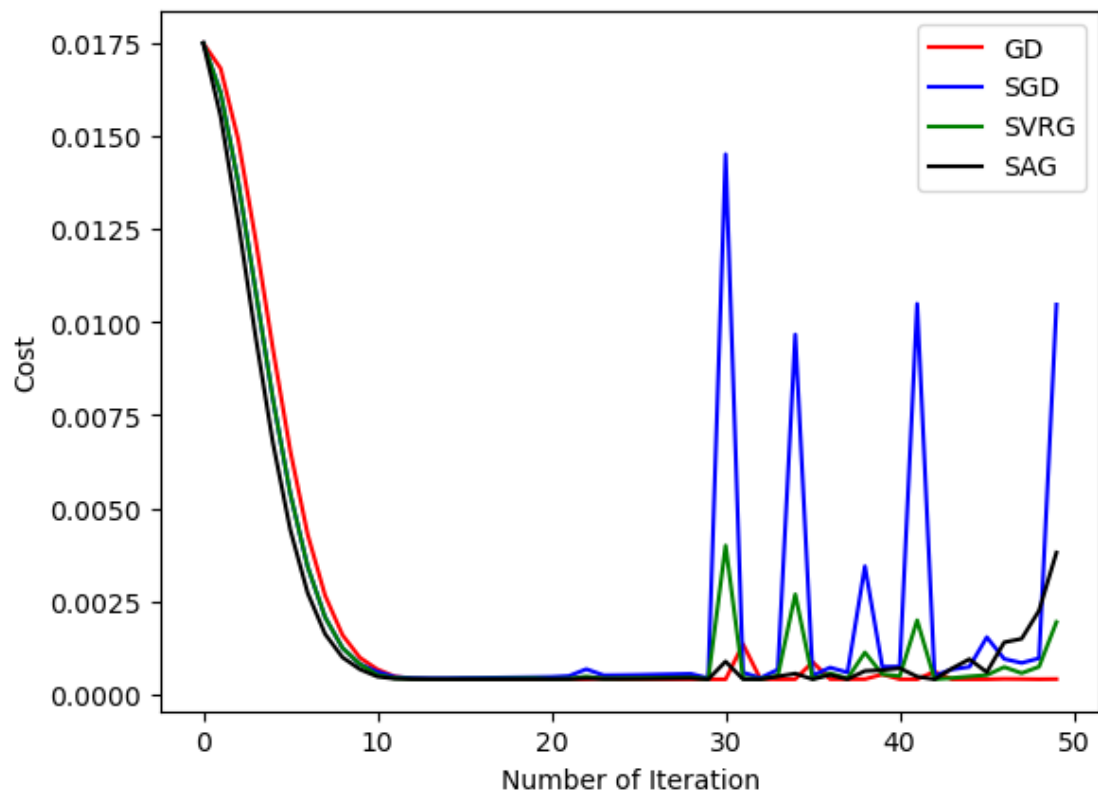
```
... 34 ...
```

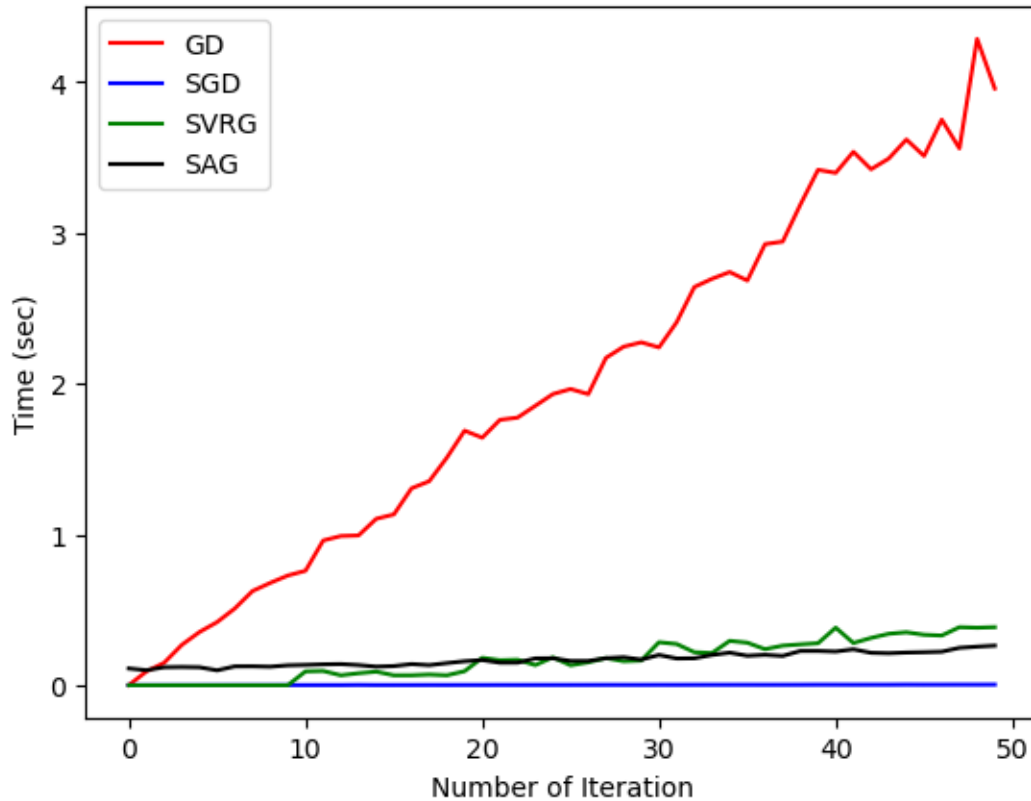
```
... 35 ...
```

```
... 36 ...
```

```
... 37 ...
```

... 38 ...
... 39 ...
... 40 ...
... 41 ...
... 42 ...
... 43 ...
... 44 ...
... 45 ...
... 46 ...
... 47 ...
... 48 ...
... 49 ...





[22]: *## PART 2: Tuning the hyper-paramter here:*

```
# Define lambda and alpha ranges
lambda_set = [0.0001, 0.001, 0.01, 0.1, 1, 10]
alpha_set= [0.0001, 0.001, 0.01, 0.1]
```

```
min_cost_gd = float('inf')
lambda_opt_gd = None
alpha_opt_gd = None
```

```
min_cost_sgd = float('inf')
lambda_opt_sgd = None
alpha_opt_sgd = None
```

```
min_cost_svrg = float('inf')
lambda_opt_svrg = None
alpha_opt_svrg = None
```

```
min_cost_sag = float('inf')
lambda_opt_sag = None
```



```

alpha_opt_sag = None

for lambda_ in lambda_set:
    for alpha in alpha_set:
        gde = solver(x, y, w_prime,alpha, num_iters, lambda_,epsilon,
↪ "GD",mem=False) # complete the command
        cost_gd = cost(x, y, gde, lambda_=lambda_)
        if cost_gd < min_cost_gd:
            min_cost_gd = cost_gd
            lambda_opt_gd = lambda_
            alpha_opt_gd = alpha

        sgde = solver(x, y, w_prime,alpha, num_iters, lambda_,epsilon,
↪ "SGD",mem=False) # complete the command
        cost_sgd = cost(x, y, sgde, lambda_=lambda_)
        if cost_sgd < min_cost_sgd:
            min_cost_sgd = cost_sgd
            lambda_opt_sgd = lambda_
            alpha_opt_sgd = alpha

        svrge = solver(x, y, w_prime,alpha, num_iters, lambda_,epsilon,
↪ "SVRG",mem=False) # complete the command
        cost_svrge = cost(x, y, svrge, lambda_=lambda_)
        if cost_svrge < min_cost_svrge:
            min_cost_svrge = cost_svrge
            lambda_opt_svrge = lambda_
            alpha_opt_svrge = alpha

        sage = solver(x, y, w_prime,alpha, num_iters, lambda_,epsilon,
↪ "SAG",mem=False) # complete the command
        cost_sag = cost(x, y, sage, lambda_=lambda_)
        if cost_sag < min_cost_sag:
            min_cost_sag = cost_sag
            lambda_opt_sage = lambda_
            alpha_opt_sage = alpha

print("The optimal lambda for GD:", lambda_opt_gd)
print("Optimal alpha for GD:", alpha_opt_gd)

print("The optimal lambda for SGD:", lambda_opt_sgd)
print("Optimal alpha for SGD:", alpha_opt_sgd)

print("The optimal lambda for SVRG:", lambda_opt_svrge)
print("Optimal alpha for SVRG:", alpha_opt_svrge)

print("The optimal lambda for SAG:", lambda_opt_sag)

```

```
print("Optimal alpha for SAG:", alpha_opt_sag)
```

```
/var/folders/l9/nl6ry63s0m3_35qt7mrjv0fh0000gn/T/ipykernel_15949/4027734134.py:9
: RuntimeWarning: overflow encountered in exp
  exp_Z = np.exp(Z)
/var/folders/l9/nl6ry63s0m3_35qt7mrjv0fh0000gn/T/ipykernel_15949/2465893598.py:1
0: RuntimeWarning: overflow encountered in exp
  value += np.log(1+np.exp(Z))

The optimal lambda for GD: 0.0001
Optimal alpha for GD: 0.1
The optimal lambda for SGD: 0.0001
Optimal alpha for SGD: 0.1
The optimal lambda for SVRG: None
Optimal alpha for SVRG: None
The optimal lambda for SAG: None
Optimal alpha for SAG: None
```

```
[23]: ## Comparing different optimizers here:
y = np.array(Y_train.iloc[0:2336])
x = np.array(X_train.iloc[0:2336,:])
N,D = x.shape
w_prime = np.random.rand(D,1)*0.01 # Initialization of w
# initialize hyperparameters
alpha = 0.1
lambda_ = 0.0001
epsilon = 1e-6

# initialize results arrays
ti = np.zeros((50,4))
cost_ = np.zeros((50,4))
mem = np.zeros((50,4))

for i in range(50):
    print(".....",i,".....")
    #-----GD-----
    start = time.time()
    gde = solver(x, y, w_prime, alpha, num_iters=i, lambda_=lambda_,
    ↪epsilon=epsilon, optimizer="GD", mem=False)
    end = time.time()
    cost_[i,0] = cost(x,y,gde,lambda_)
    ti[i,0] = end-start
    process = psutil.Process()
    mem[i,0] = process.memory_info().rss / 1024 / 1024 # memory usage in MB

    #-----SGD-----
    #Complete for SGD solver here :
    start = time.time()
```

```

    sgde = solver(x, y, w_prime, alpha, num_iters=i, lambda_=lambda_,
↪epsilon=epsilon, optimizer="SGD", mem=False)
    end = time.time()
    cost_[i,1] = cost(x,y,sgde,lambda_)
    ti[i,1] = end-start
    process = psutil.Process()
    mem[i,1] = process.memory_info().rss / 1024 / 1024

    #-----SVRG-----
    #Complete for SVRG solver here :
    start = time.time()
    svrge = solver(x,y, w_prime, alpha, num_iters=i, lambda_=lambda_,
↪epsilon=epsilon, optimizer = "SVRG",mem=False)
    end = time.time()
    cost_[i,2] = cost(x,y,svrge,lambda_)
    ti[i,2] = end-start
    process = psutil.Process()
    mem[i,2] = process.memory_info().rss / 1024 / 1024

    #-----SAG-----
    #Complete for SAG solver here :
    start = time.time()
    sage = solver(x, y, w_prime, alpha, num_iters=i, lambda_=lambda_,
↪epsilon=epsilon, optimizer = "SAG",mem=False)
    end = time.time()
    cost_[i,3] = cost(x,y,sage,lambda_)
    ti[i,3] = end-start
    process = psutil.Process()
    mem[i,3] = process.memory_info().rss / 1024 / 1024

## Pl the results:
10 = plt.plot(cost_[:,0],color="red")
# complete other plots here:
11 = plt.plot(cost_[:,1],color="blue")
12 = plt.plot(cost_[:,2],color="green")
13 = plt.plot(cost_[:,3],color="black")

plt.xlabel("Number of Iteration")
plt.ylabel("Cost")
plt.legend(['GD', 'SGD', 'SVRG', 'SAG'])
plt.show()

10 = plt.plot(ti[:,0],color="red")

```

```

# complete other plots here:
l1 = plt.plot(ti[:,1],color="blue")
l2 = plt.plot(ti[:,2],color="green")
l3 = plt.plot(ti[:,3],color="black")

plt.xlabel("Number of Iteration")
plt.ylabel("Time (sec)")
plt.legend(['GD', 'SGD', 'SVRG', 'SAG'])
plt.show()

l0 = plt.plot(mem[:,0],color="red")
# complete other plots here:
l1 = plt.plot(mem[:,1],color="blue")
l2 = plt.plot(mem[:,2],color="green")
l3 = plt.plot(mem[:,3],color="black")

plt.xlabel("Number of Iteration")
plt.ylabel("Memory (MB)")
plt.legend(['GD', 'SGD', 'SVRG', 'SAG'])
plt.show()

```

```
... 0 ...
```

```

/var/folders/l9/nl6ry63s0m3_35qt7mrjv0fh0000gn/T/ipykernel_15949/4027734134.py:9
: RuntimeWarning: overflow encountered in exp
  exp_Z = np.exp(Z)

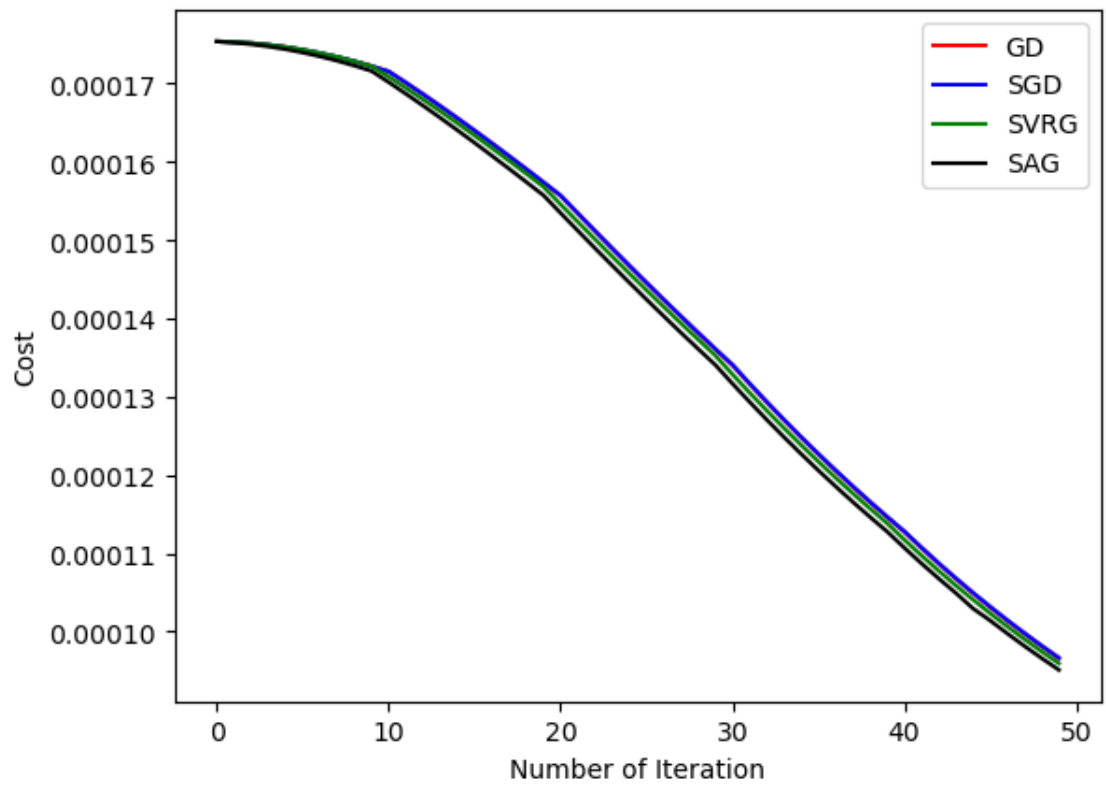
```

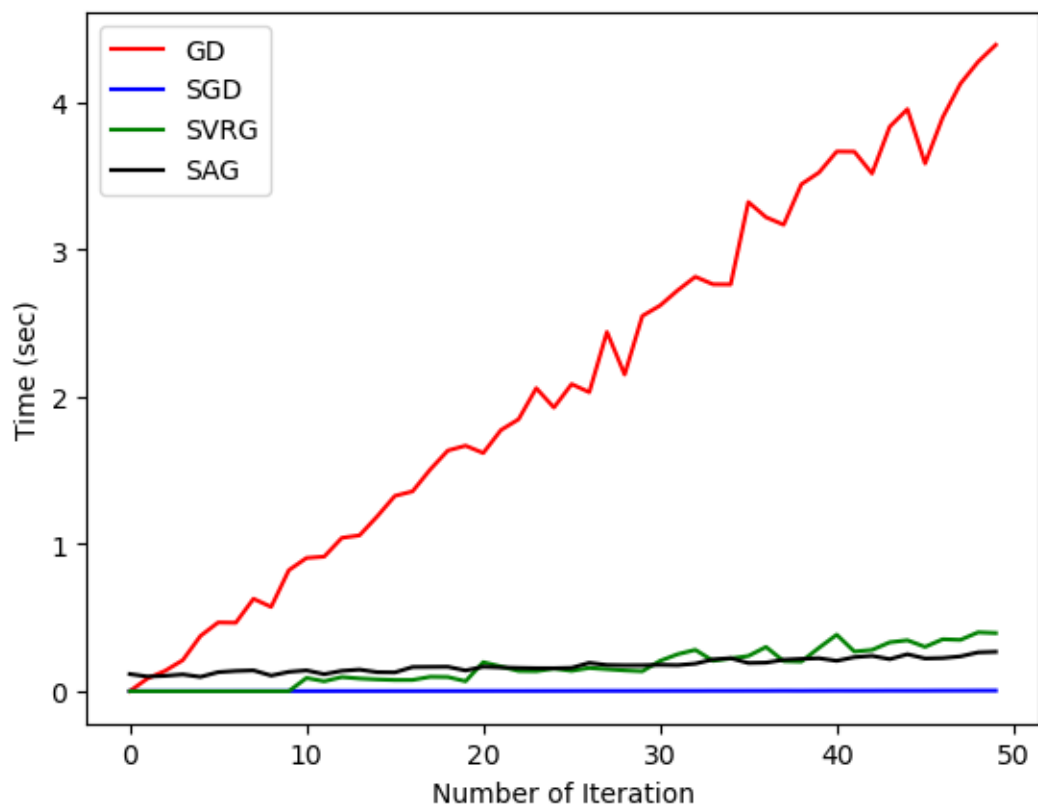
```

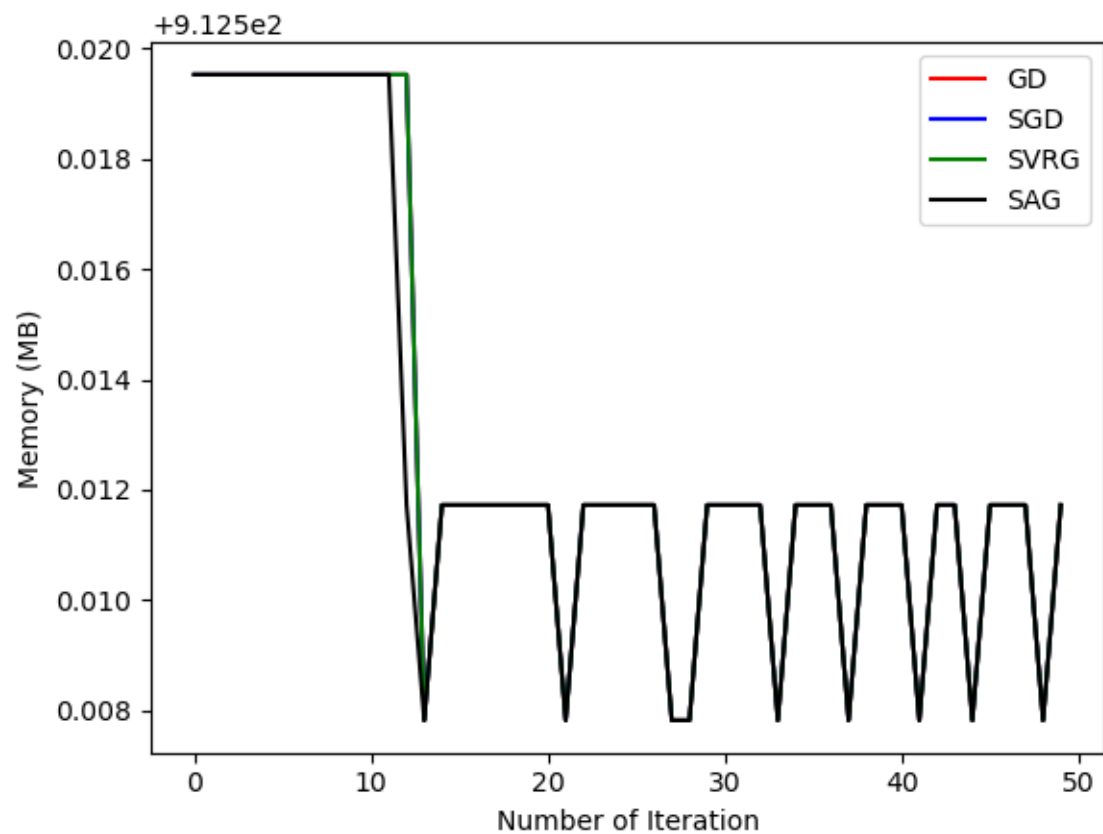
... 1 ...
... 2 ...
... 3 ...
... 4 ...
... 5 ...
... 6 ...
... 7 ...
... 8 ...
... 9 ...
... 10 ...
... 11 ...
... 12 ...
... 13 ...
... 14 ...
... 15 ...
... 16 ...
... 17 ...
... 18 ...
... 19 ...
... 20 ...
... 21 ...
... 22 ...

```

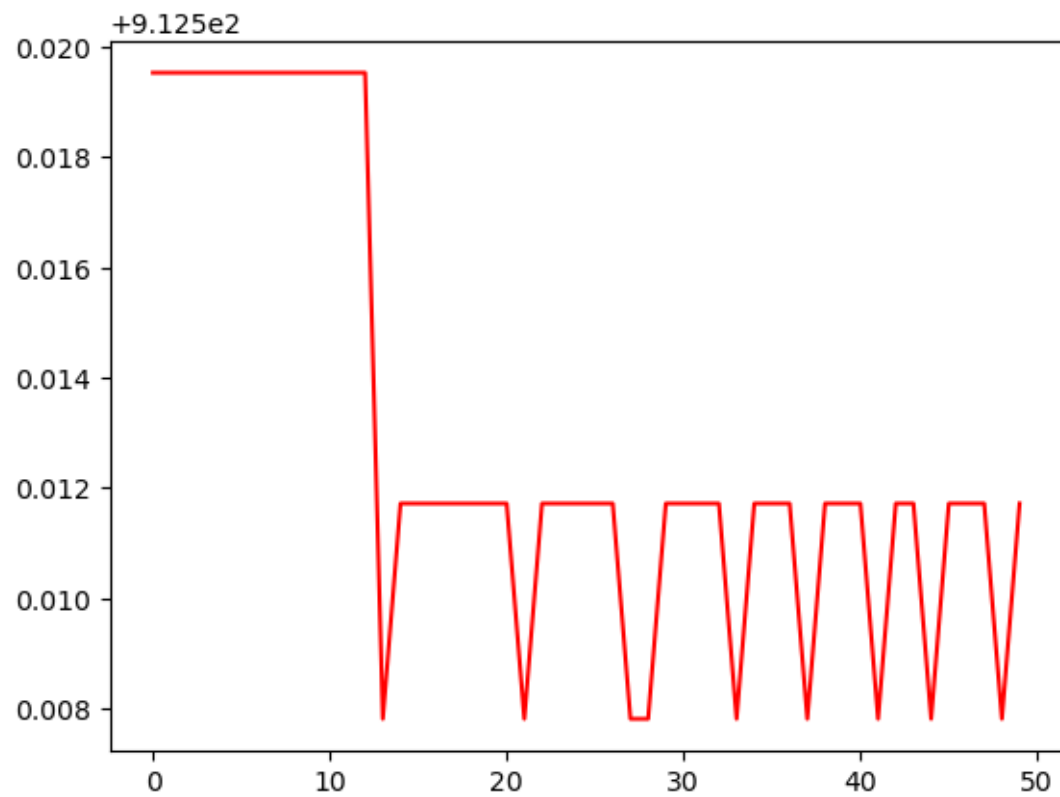
... 23 ...
... 24 ...
... 25 ...
... 26 ...
... 27 ...
... 28 ...
... 29 ...
... 30 ...
... 31 ...
... 32 ...
... 33 ...
... 34 ...
... 35 ...
... 36 ...
... 37 ...
... 38 ...
... 39 ...
... 40 ...
... 41 ...
... 42 ...
... 43 ...
... 44 ...
... 45 ...
... 46 ...
... 47 ...
... 48 ...
... 49 ...



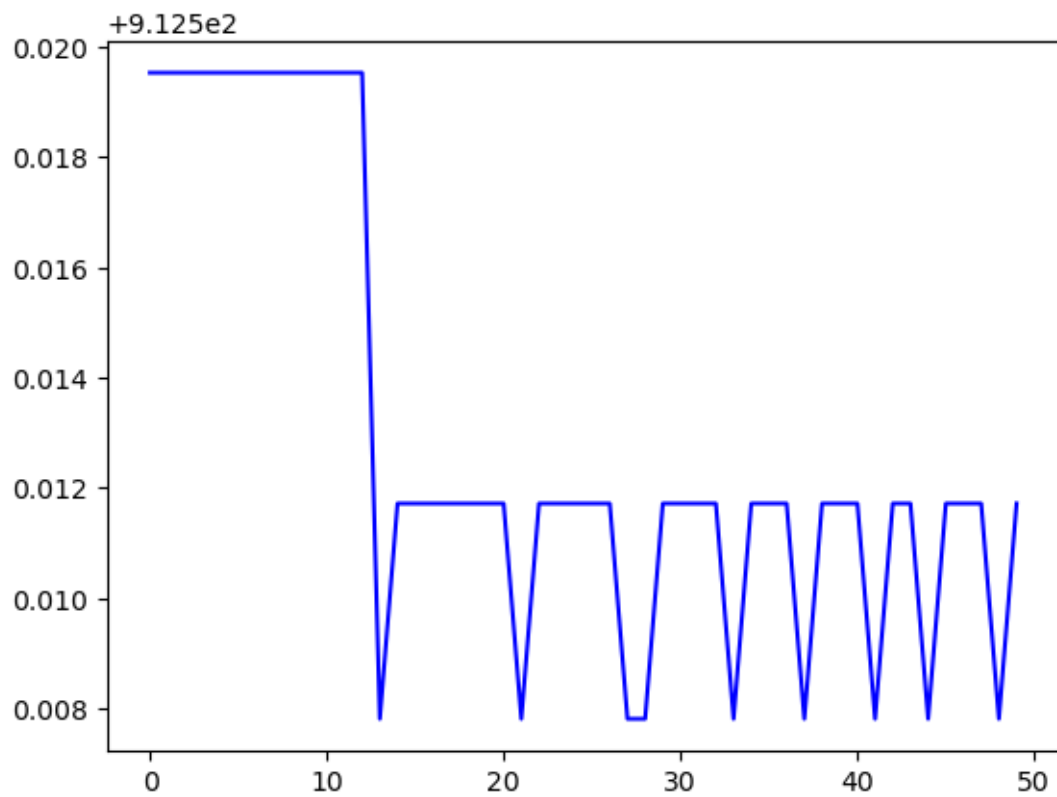




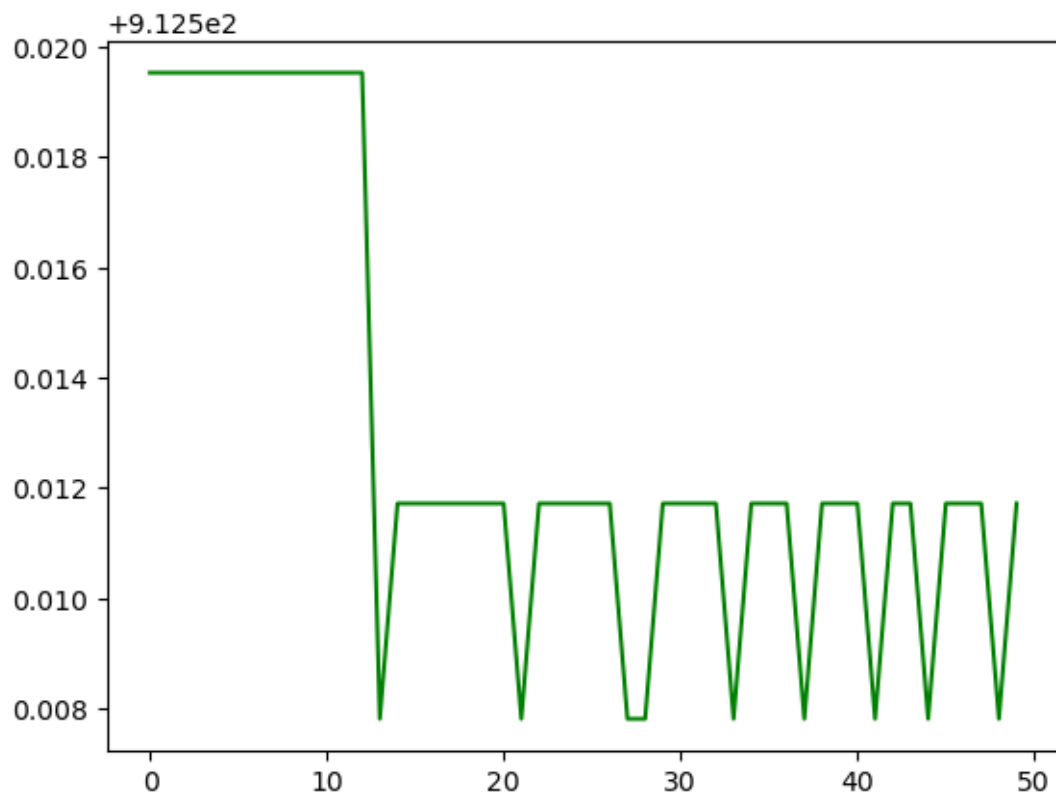
```
[24]: 10 = plt.plot(mem[:,0],color="red")
plt.show()
```

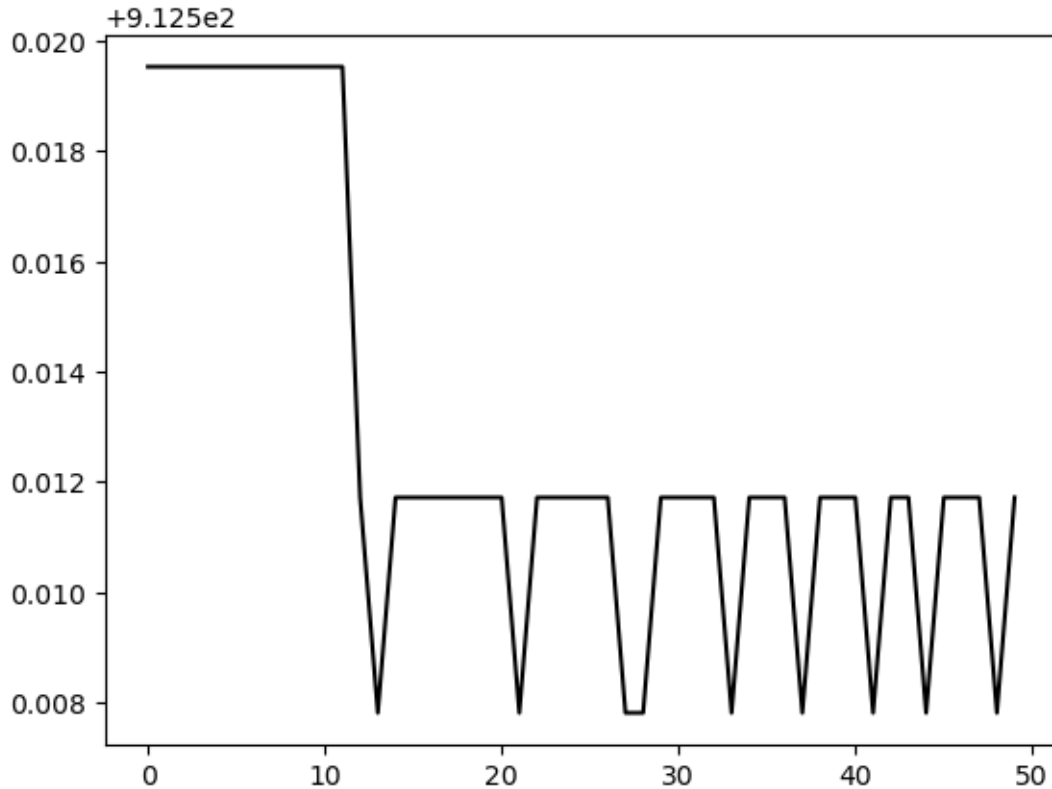
```
[25]: l1 = plt.plot(mem[:,1],color="blue")  
plt.show()
```



```
[26]: 12 = plt.plot(mem[:,2],color="green")  
plt.show()
```



```
[27]: 13 = plt.plot(mem[:,3],color="black")  
plt.show()
```



```
[28]: gde = solver(x, y, w_prime, alpha, num_iters=i, lambda_=lambda_,
    ↪epsilon=epsilon, optimizer="GD", mem=True)
sgde = solver(x, y, w_prime, alpha, num_iters=i, lambda_=lambda_,
    ↪epsilon=epsilon, optimizer="SGD", mem=True)
svrge = solver(x,y, w_prime, alpha, num_iters=i, lambda_=lambda_,
    ↪epsilon=epsilon, optimizer = "SVRG",mem=True)
sage = solver(x, y, w_prime, alpha, num_iters=i, lambda_=lambda_,
    ↪epsilon=epsilon, optimizer = "SAG",mem=True)
```

```
/var/folders/l9/nl6ry63s0m3_35qt7mrjv0fh0000gn/T/ipykernel_15949/4027734134.py:9
: RuntimeWarning: overflow encountered in exp
exp_Z = np.exp(Z)
```

```
mem for GD (MB): 3929190.432768
mem for GD (MB): 3929190.432768
mem for GD (MB): 3929190.432768
mem for GD (MB): 3929190.432768
mem for GD (MB): 3929190.432768
mem for SGD (MB): 3929190.432768
mem for SGD (MB): 3929190.432768
mem for SGD (MB): 3929190.432768
```

```
mem for SGD (MB): 3929190.432768
mem for SGD (MB): 3929190.432768
mem for SAG (MB): 3929190.432768
mem for SAG (MB): 3929190.432768
mem for SAG (MB): 3929190.432768
mem for SAG (MB): 3929190.432768
mem for SAG (MB): 3929190.432768
```

[]: